

Experiences in the Practice of Design of Experiments at NASA

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National Aeronautics and Space Administration

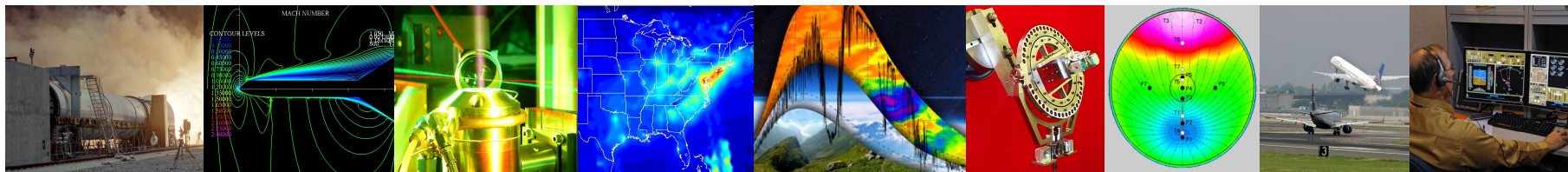
Langley Research Center

Hampton, Virginia

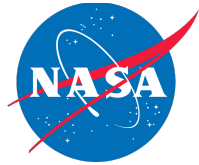
March 16, 2022

Arizona State University

Design of Experiments Course Guest Lecture



NASA Langley Research Center



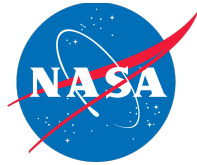
- **One of 10 NASA Centers located in Southeastern Virginia**
- **Established in 1917 as first civilian aeronautics laboratory in the U.S.**
- **Initial home of the first astronauts, Mercury 7**

Today, we focus our research on technical challenges in:

- **Space Exploration**
- **Aeronautics and Air Transportation**
- **Earth and Planetary Science**

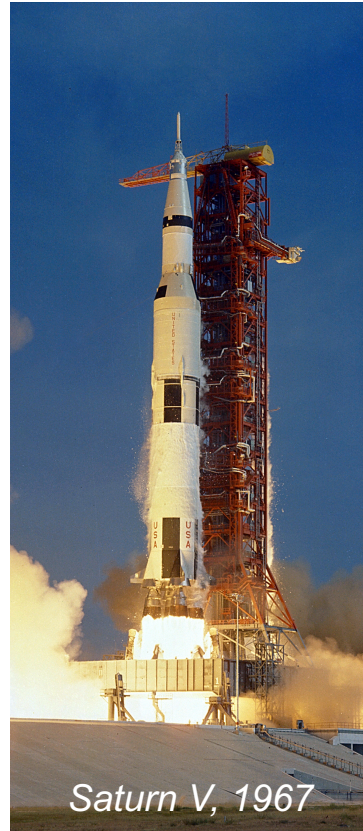


Aerospace and Statistical Timeline



Wright Flyer, 1903

NASA Formed, 1958



Saturn V, 1967

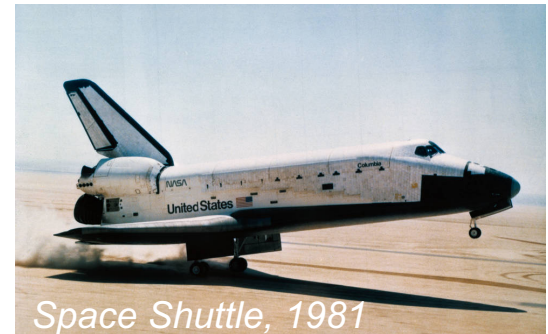


Apollo 11, 1969

“...most of the statistical work is performed by engineers and scientists, some well trained in statistics and others having only a passing acquaintance with the subject...”¹



Variable Density Tunnel, 1921



Space Shuttle, 1981

Fisher, DOE, 1920's



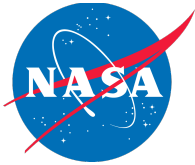
Box, RSM, 1951



Deming, Taguchi, Quality, 1980's

¹ Rubin, E. (1966) “Some Statistical Applications in the Apollo Program.” *The American Statistician*, 20 (4), pp. 32-34.

An Engineer's Statistical Awakening



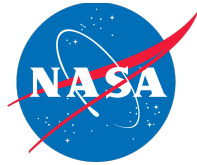
- **Traditionally, engineers lacked relevant, applicable, practical courses in statistics (*that's me*)**
 - **Engineers are generally trained to “fear” uncertainty**
- **In 1999, I stumbled upon a 1980 NASA report written by a visiting statistics professor with “radical” ideas that promised significant schedule and cost benefits.**

A few of those “radical” statistical concepts

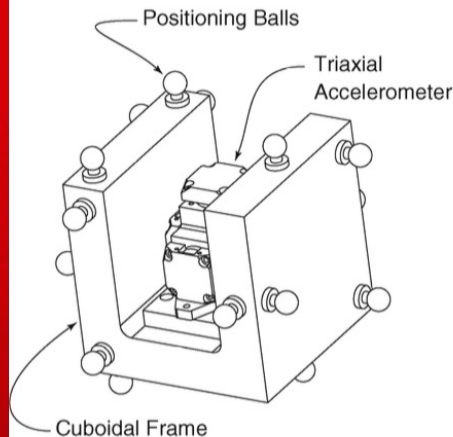
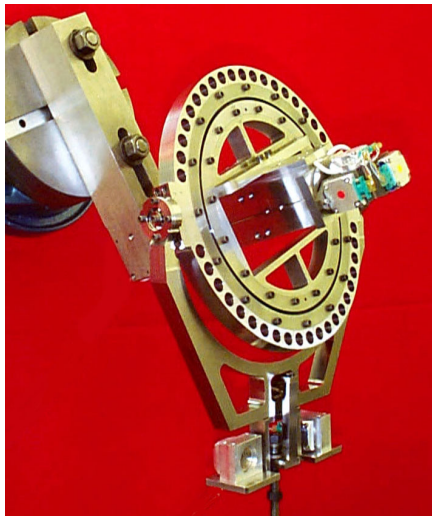
- **Experimental design starts with questions, not data**
- **Factorial experiments are efficient, not chaotic**
- **Modeling involves estimation, more than plotting data**

**Discovered “new” powerful statistical techniques to
efficiently learn and gain deeper insights,
Design of Experiments (DOE)**

Measurement System Characterization



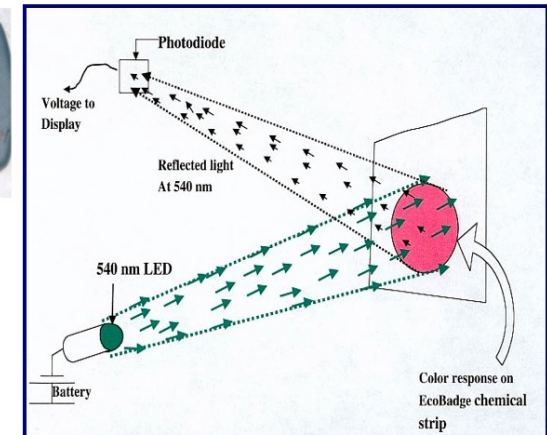
Improving aerospace research measurement systems quality



Parker, Morton, Draper, Line (2001), "A Single-Vector Force Calibration Method Featuring the Modern Design of Experiments," *AIAA 2001-0170*.

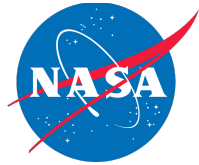
Parker and Finley (2007), "Advancements in Aircraft Model Force and Attitude Instrumentation by Integrating Statistical Methods," *AIAA Journal of Aircraft*.

Increasing accessibility and participation in climate monitoring education



Pippin, et al. (2007), "Improvements to the Passive Ozone Measurement System Used by GLOBE Schools," *American Geophysical Union Annual Conference*.

First Agency-Level DOE Demonstration



Shuttle



Ares I



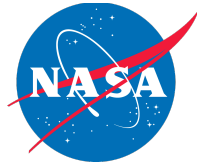
Flight Test



**Full-Scale
Static Firing**



Aerospace Wind Tunnel Testing



Characterizing aerodynamic performance of complex aircraft configurations

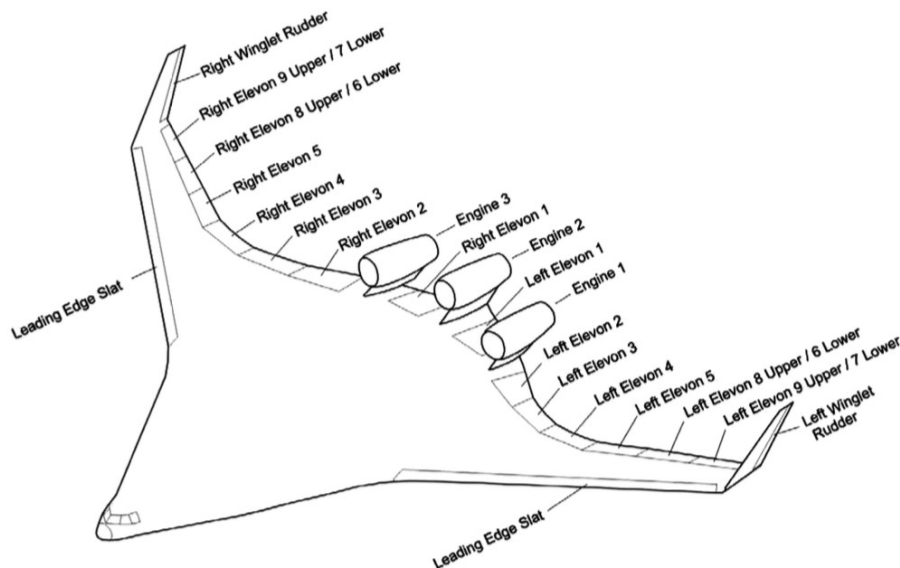


Fig. 2 BWB configuration control surfaces.

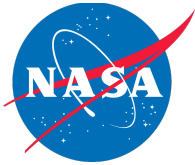
Landman, et al. (2007), "Response Surface Methods for Efficient Complex Aircraft Configuration Aerodynamic Characterization," *Journal of Aircraft*, 44, (4)

Powered test article of rotorcraft for computational model validation



Overmeyer, et al. (2015), "Case Studies for the Statistical Design of Experiments Applied to Powered Rotor Wind Tunnel Tests," AIAA 2015-2713

Practicing Design of Experiments Starts with Questions, not Data

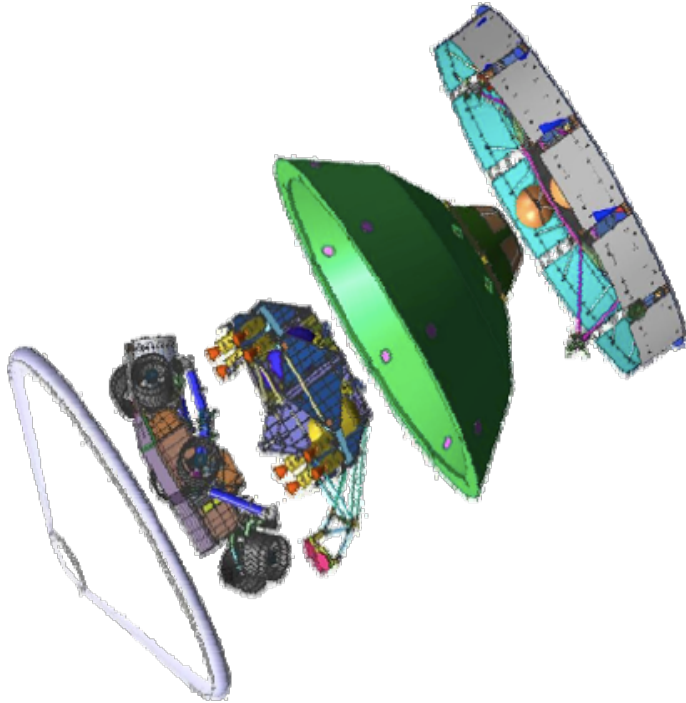
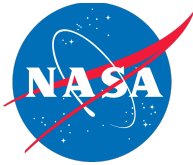


Heilmeier's preflight checklist for successfully launching a project
“to curb and clarify both the enthusiasm of the researchers and to
evaluate the resource demands of the project managers”

- **What are we seeking to learn?**
 - Are the objectives quantifiable, detectable, measurable?
 - What is the impact if you we successful?
- **How well do we need to know?**
 - How much risk are we willing to accept in being wrong?
 - What are the consequences if we are wrong?
- **Are the resource demands defensible to meet our objectives?**

**Questions apply recursively throughout project life-cycle, and
vertically from component testing up to vehicle design.**

Mars Planetary Entry, Descent, and Landing

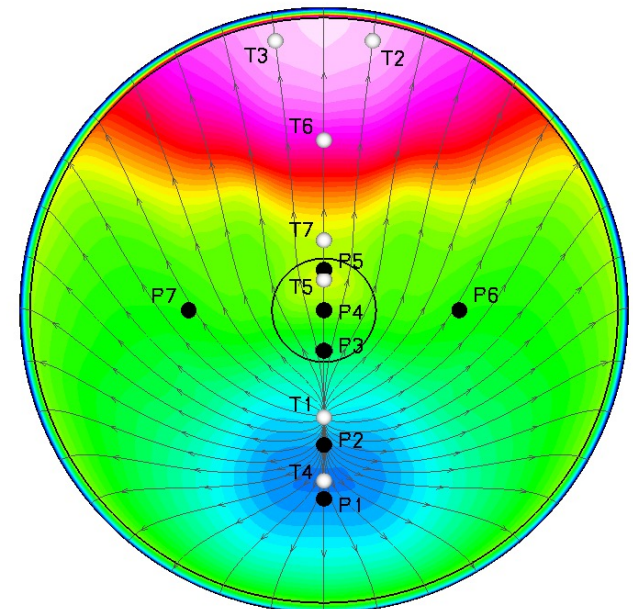


Objective: Reduce uncertainty in pre-flight landing ellipse estimation

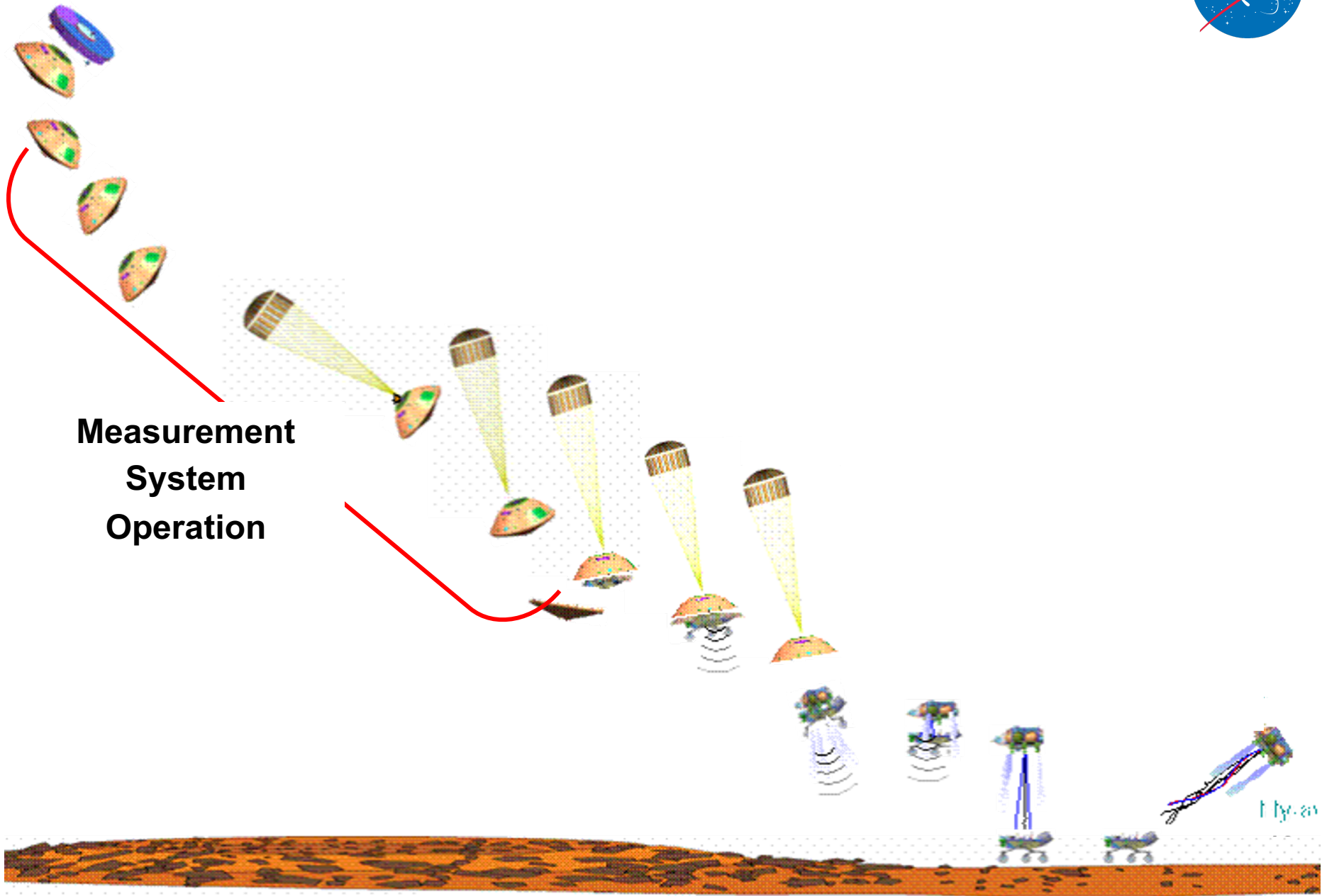
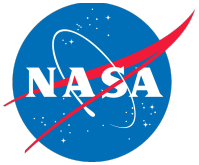
Develop a measurement system to enable trajectory reconstruction of Mars entry to refine computational models

Measurement System Characterization

- **System of pressure sensors and electronics**
- **Pressures vary spatially across the heat shield and temporally throughout entry**
- **Temperatures conditions vary at start of entry and across the heatshield during entry**



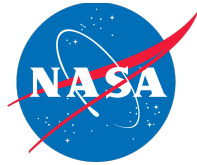
Entry, Descent, and Landing Timeline



**Measurement
System
Operation**

Fly-av

Experimental Design Factors

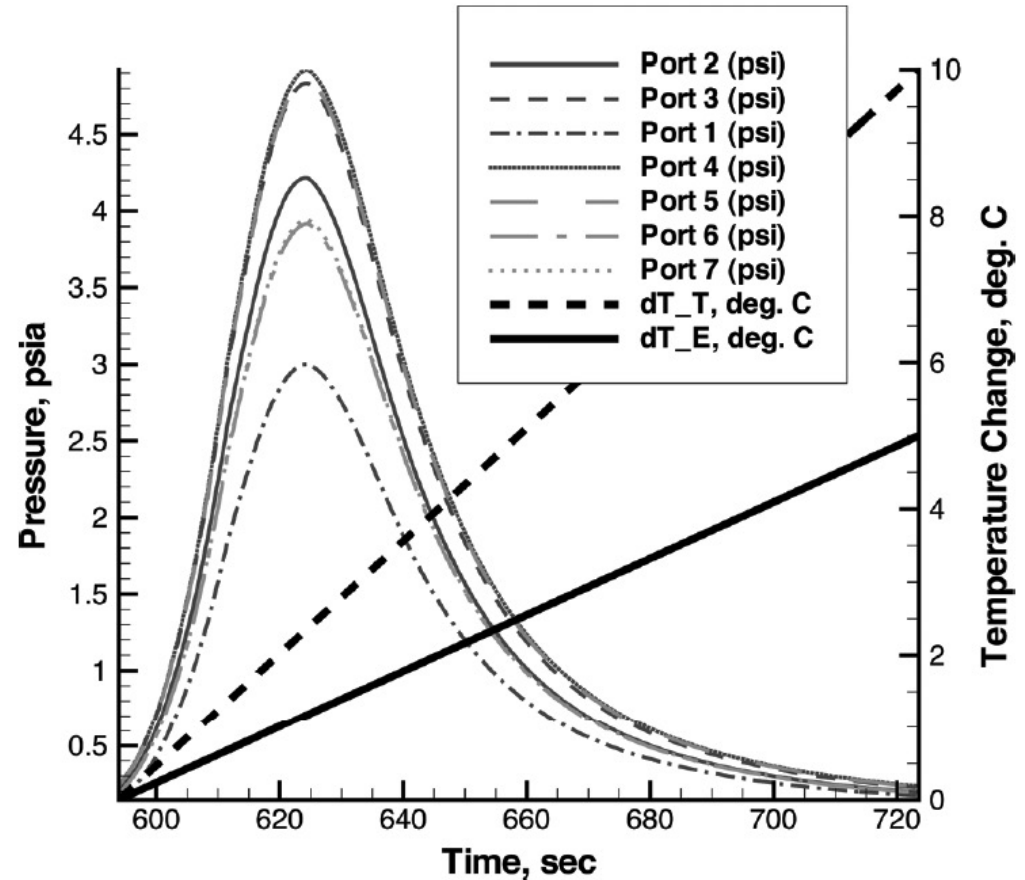


Characterization Design Space

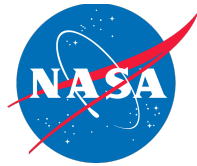
Pressure	0 to 5 psia
Temp. Sensor	-120 to -60 deg.C
Temp. SSE	-20 to +55 deg.C

Sensor – multiple pressure transducers

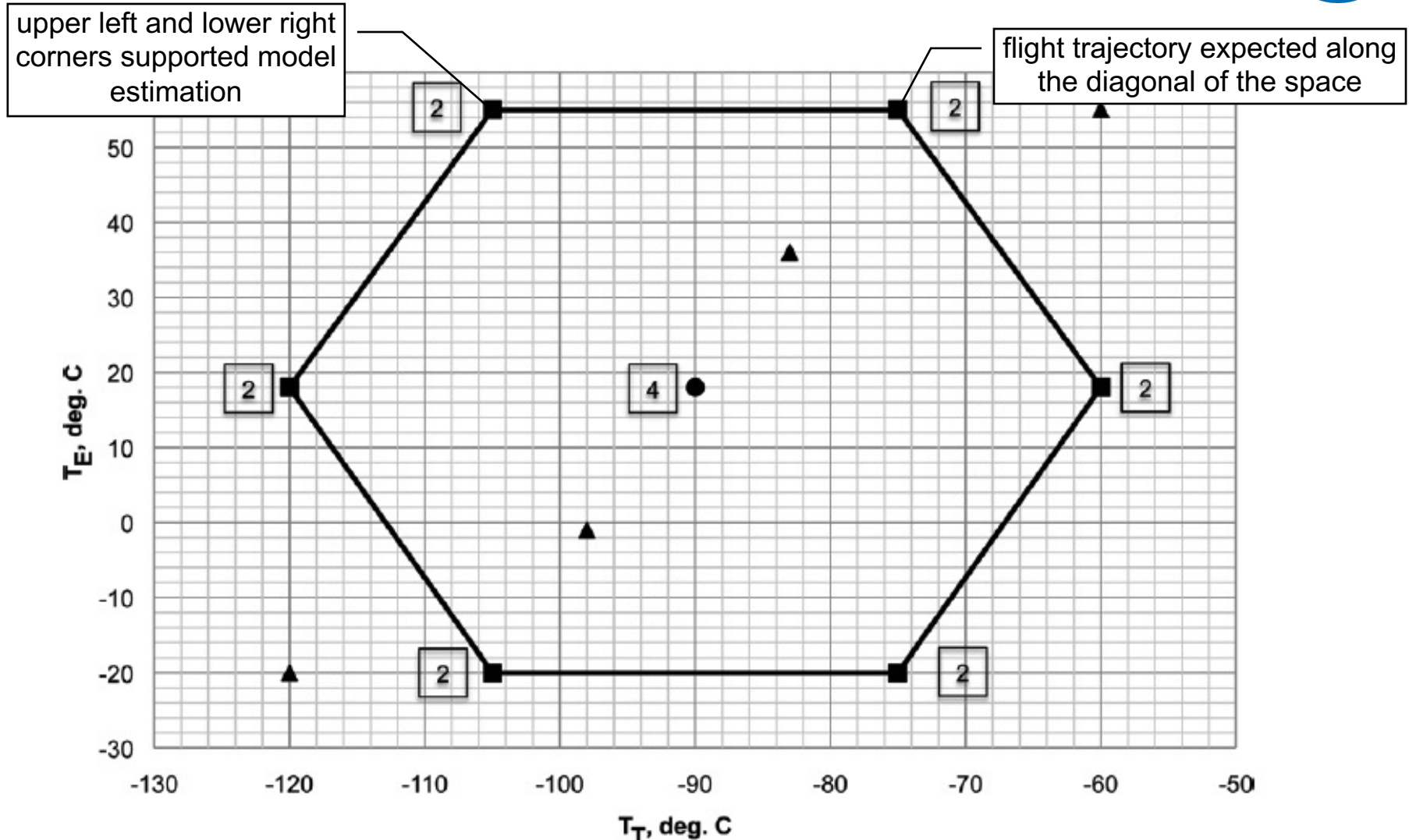
SSE – signal support electronics form the data acquisition system (DAS)



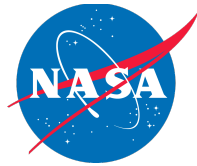
- Temperatures vary between DAS and sensor locations (start and entry)
- Pressure varies across port locations



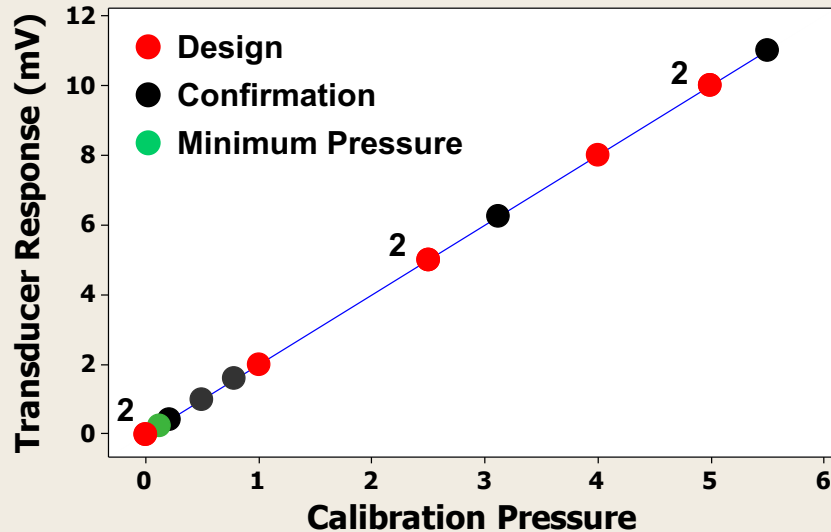
Temperature Design Space



- Superimposed designs to support multiple objectives
 - Model estimation, Confirmation, and Flight Trajectory



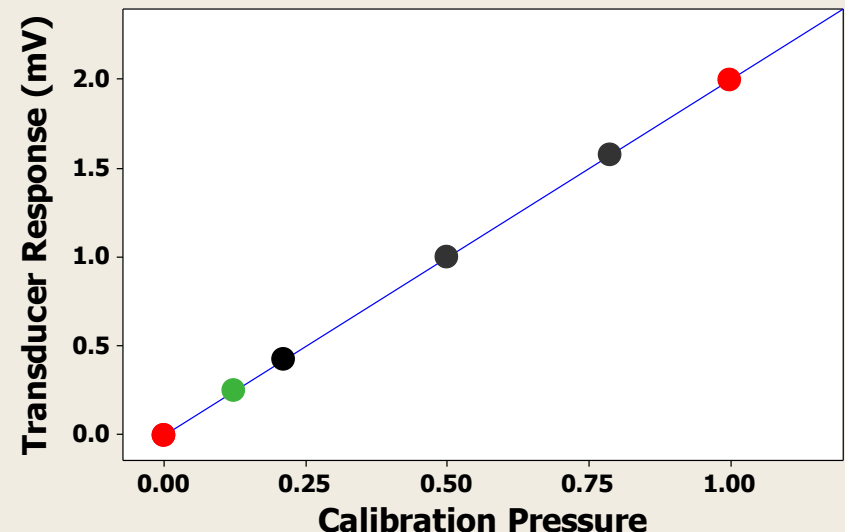
Pressure Design Space



Allocation of Design Points

- 0-5 psia calibration (8)
- 0-1 psia low-end calib./conf. (3)
- 0.12 psia confirmation (1)
- 5.5 psia max confirmation (1)
- random conf. at mid-range (1)

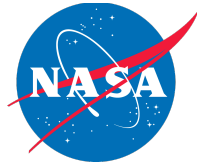
Inscribed Design within 0 to 1 psia



Split-Plot Design, Nested randomization

- Once a temperature combination is set, pressure settings are completely randomized

Numbers indicate replicated design points



Mathematical Model

ability to update and estimate in flight

Primary Sensitivity Coefficient

zero intercept adjustments as a function of temperature

$$V = \beta_0 + \beta_1 P + \beta_2 T_{sensor} + \beta_3 T_{SSE}$$

second-order effect of pressure (non-linearity)

$$+ \beta_{11} P^2 + \beta_{22} T_{sensor}^2 + \beta_{33} T_{SSE}^2$$

second-order effects of temperature on intercept

$$+ \beta_{12} P \times T_{sensor} + \beta_{13} P \times T_{SSE}$$

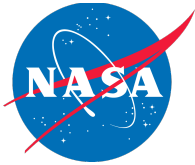
sensitivity adjustments as a function of temperature

$$+ \beta_{23} T_{sensor} \times T_{SSE} + \varepsilon$$

assumed negligible based on system knowledge

Interpretation and Translation into the Subject-Matter Experts' Nomenclature is Vital

Model Development and Application



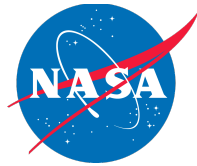
**Characterization Model Developed on Earth in
Environmental Simulation Laboratory**

$$V = f(P, T_{sensor}, T_{SSE})$$

**Inverse Relationship used to Estimate
Pressure During Mars Entry**

$$\hat{P} = f(V, T_{sensor}, T_{SSE})$$

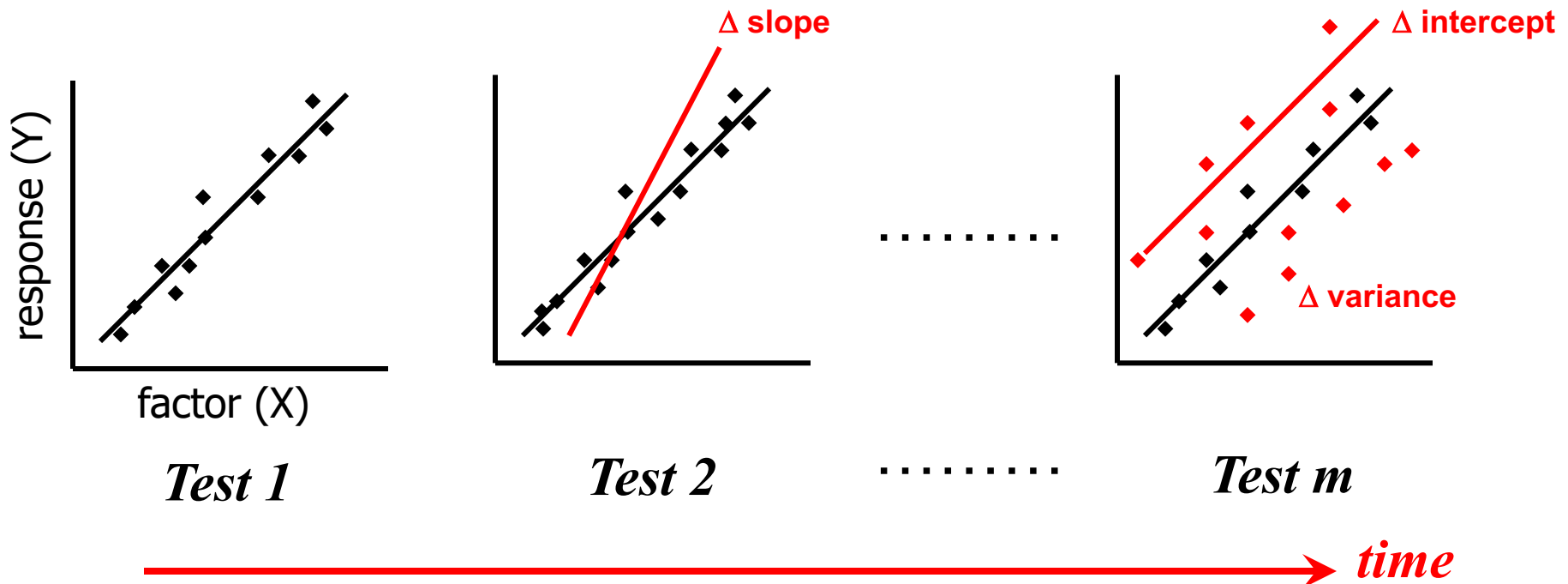
**Inverse Model includes Estimated Uncertainty that
Influences Flight Reconstruction Fidelity**



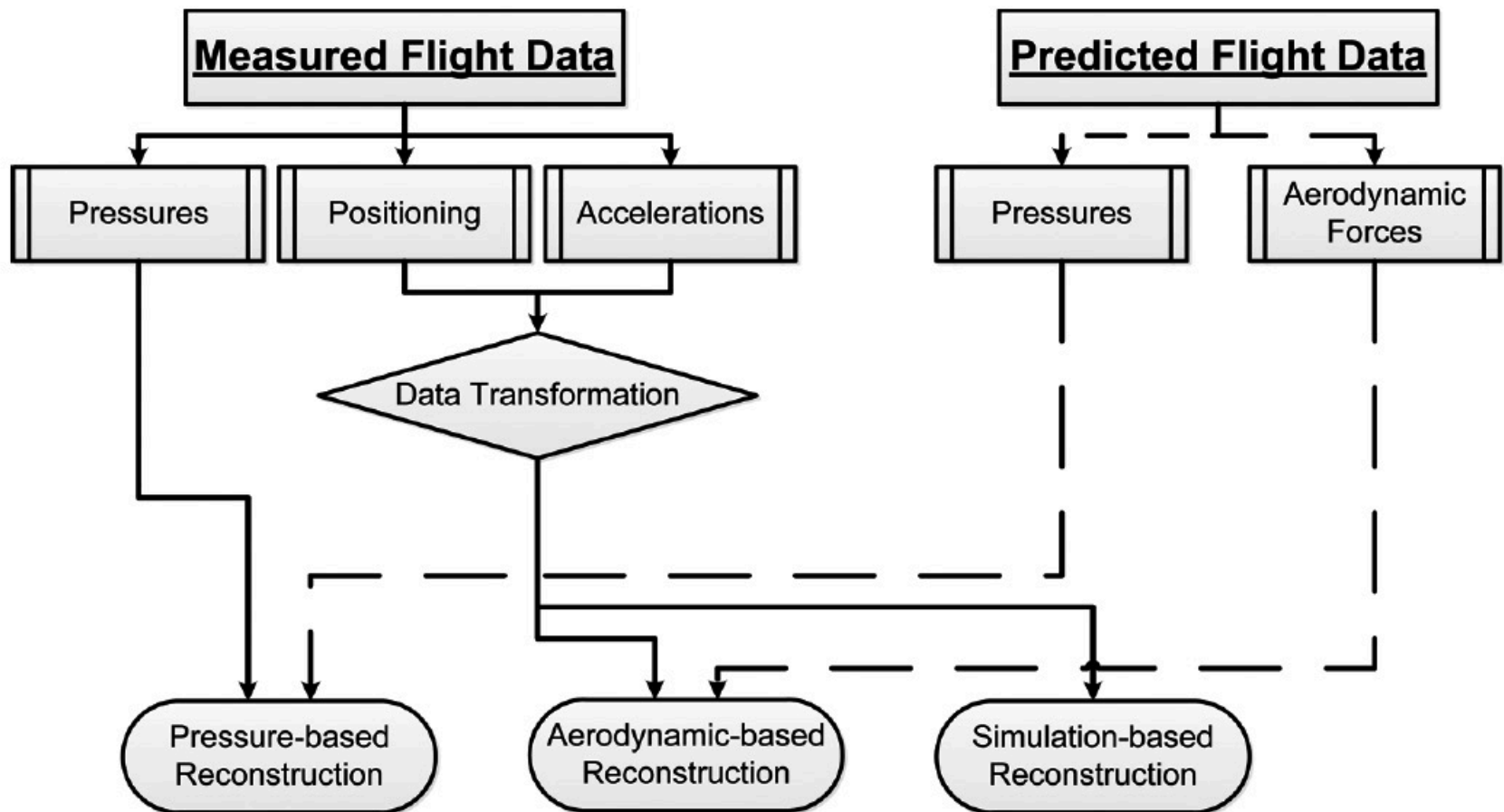
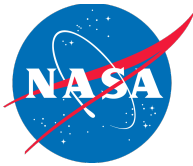
Calibration Model Stability

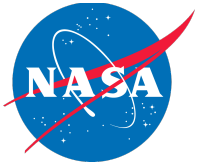
- Stability tests are performed periodically to monitor the calibration model coefficients, not the raw data

$$V = \beta_0 + \beta_1 P + \beta_2 T_{\text{sensor}} + \beta_3 T_{\text{SSE}} + \beta_{11} P^2 + \beta_{22} T_{\text{sensor}}^2 + \beta_{33} T_{\text{SSE}}^2 \\ + \beta_{12} P \times T_{\text{sensor}} + \beta_{13} P \times T_{\text{SSE}} + \beta_{23} T_{\text{sensor}} \times T_{\text{SSE}} + \varepsilon$$



Models Integrated into Flight Reconstruction





Concluding Remarks

- **The practice of DOE relies on effective multi-disciplinary collaboration and the art of translating statistical concepts into subject-matter vernacular**
- **Infusing statistical DOE at NASA has influenced aerospace research and development, over the past 25 years**
- **Successful demonstrations have resulted in broader adoption of DOE philosophy and depth in DOE discipline expertise**